



Eidgenössische Technische Hochschule Zürich
Swiss Federal Institute of Technology Zurich

Relating compound toxicity to molecular structure using machine learning

Master Thesis

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Abstract

Abstract goes here.

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Chapter 1

Introduction

intro

Chapter 2

Literature Review

2.1 Background

2.2 Context

Material and Methods

3.1 Dataset

Consider a collection of m assay endpoints, denoted by $A = \{a_1, a_2, \dots, a_m\}$ and a set of n compounds represented as $C = \{c_1, c_2, \dots, c_n\}$. We introduce a *presence matrix* $P \in \{0, 1\}^{m \times n}$. In this matrix, each row, indexed by i , corresponds to an individual assay endpoint a_i , and each column, indexed by j , signifies the presence (1) or absence (0) of a compound c_j in the respective assay endpoints. For a visual representation, refer to Figure 3.1, which illustrates the *presence matrix* P encompassing all assay endpoints and compounds available in the *invitroDBv3.5* dataset. A compound is considered present in an assay endpoint if it has undergone testing, leading to the availability of a corresponding concentration-response series. The sparsity of matrix P arises from the fact that not all compounds undergo testing across all assay endpoints.

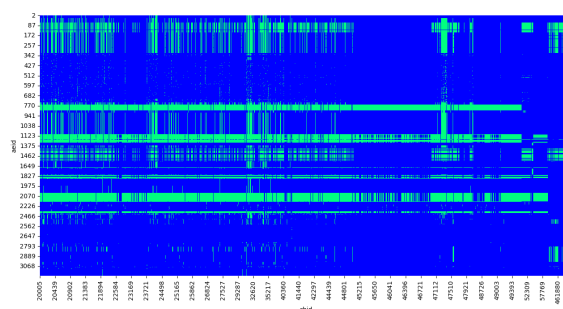


Figure 3.1: The *presence matrix* P for $m = 271$ assay endpoints and $n = 9000$ compounds. The count, where $P_{ij} = 1$, indicates the availability of 3M concentration-response series for downstream analysis.

A *concentration-response series* is represented as a set of k_{ij} concentration-response pairs:

$$S = \{(conc_1, resp_1), (conc_2, resp_2), \dots, (conc_k, resp_k)\}$$

where $conc_i$ values are not necessarily unique. In practice, concentrations are often subjected to multiple testing iterations, resulting in the formation of n_{conc} distinct concentration groups. Within each concentration group, the number of replicates is indicated by n_{rep} . Concentrations are transformed to the logarithmic scale using the unit μM (micromolar), while the responses are normalized to either fold-induction or percent-of-control units. Figure 3.2 showcases a concentration-response series for a compound tested within a single assay endpoint.

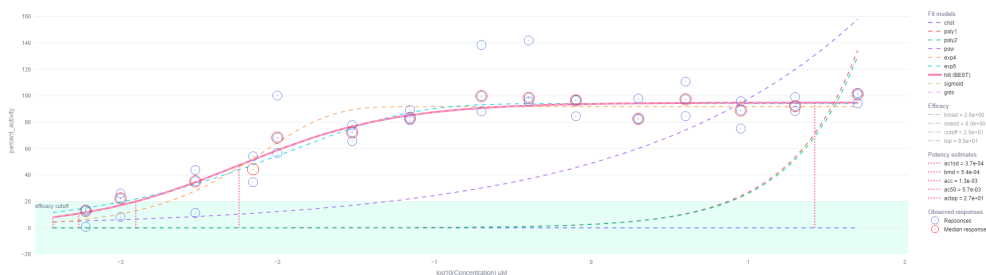


Figure 3.2: A concentration-response series for the compound *Estropipate* in the assay endpoint *TOX21_ERa_LUC_VM7_Agonist*. The series has a total of $k = 45$ concentration-response pairs and is composed of $n_{conc} = 15$ concentration groups, each with $n_{rep} = 3$ replicates.

3.2 Pytcpl

We introduce **pytcpl**, a streamlined Python package inspired by the R package **tcpl**, designed for processing high-throughput screening data. Our package is crafted to accomodate customizable processing steps and facilitate interactive data visualization with **curve surfer** and empowers Python-oriented researchers to seamlessly engage in data analysis and exploration. It primarily focuses on providing essential features such as concentration-response curve fitting and allows for continuous hit-calling for compound bioactivity across diverse assay endpoints, akin to **tcplfit2**. Optionally, the **Invitrodb version 3.5 release** can serve as backend database if desired. The package optimizes data storage and compresses raw data and metadata from *invitroDB* into Parquet files. This efficient strategy reduces storage needs, resulting in just 4 GB within the repository—compared to the original 80 GB database.

This obviates the need for a cumbersome, large-scale database installation, rendering downstream analysis more accessible and efficient.

3.3 Machine Learning Pipeline

Chapter 4

Results and Discussion

sectionResults sectionEvaluation sectionDiscussion

Chapter 5

Conclusion

Appendix A

Appendix



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